

# Robust 3D Object Detection using Probabilistic Point Clouds from Single Photon LiDARs

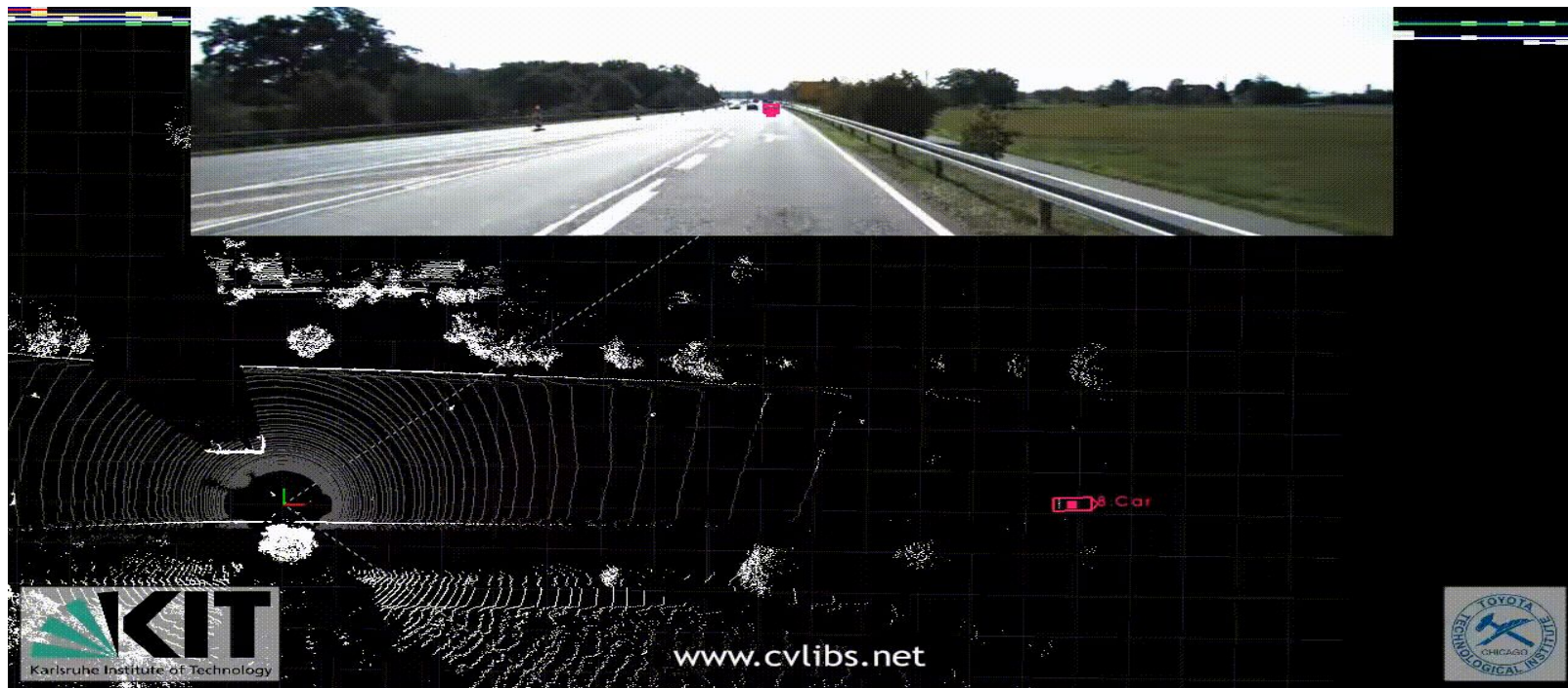
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ICCV 2025



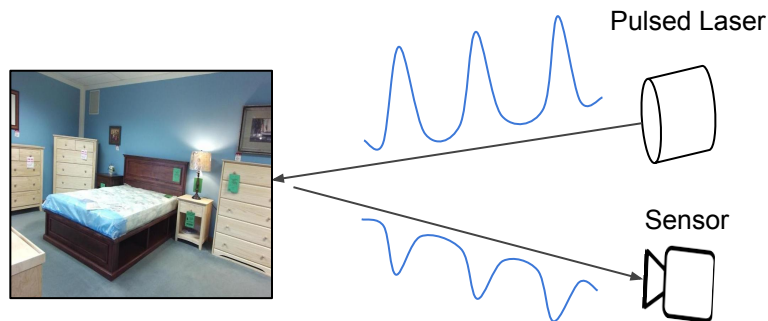
<https://bhavyagoyal.github.io/ppc>

# 3D Inference with LiDAR Point Clouds

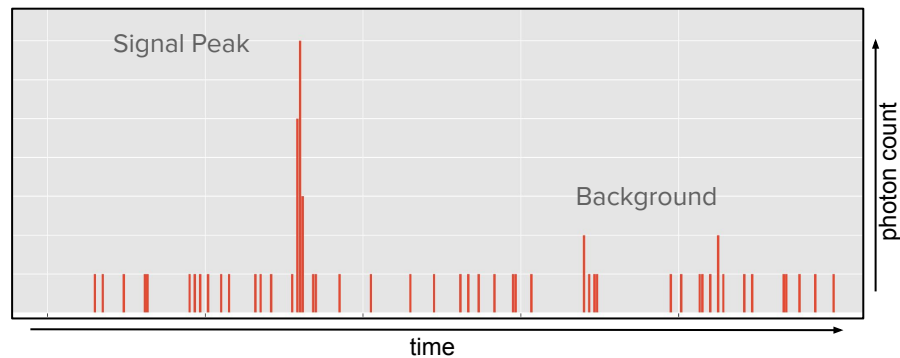


Point Clouds are often **Sparse** and **Noisy**  
**Smaller** and **Farther** objects are challenging

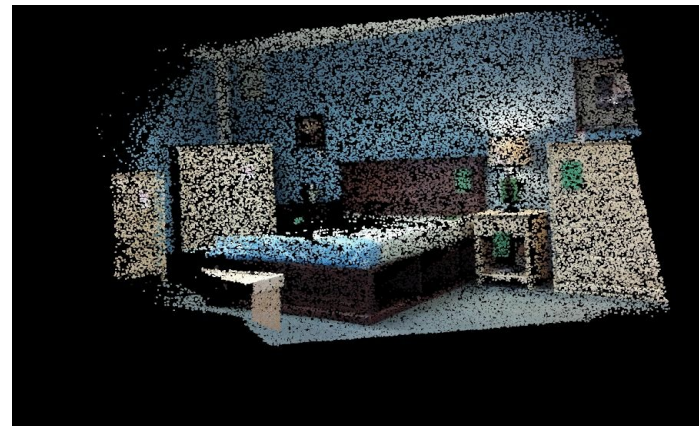
# 3D Sensing with a LiDAR



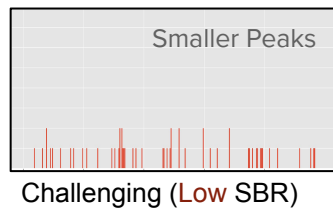
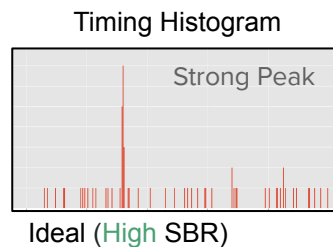
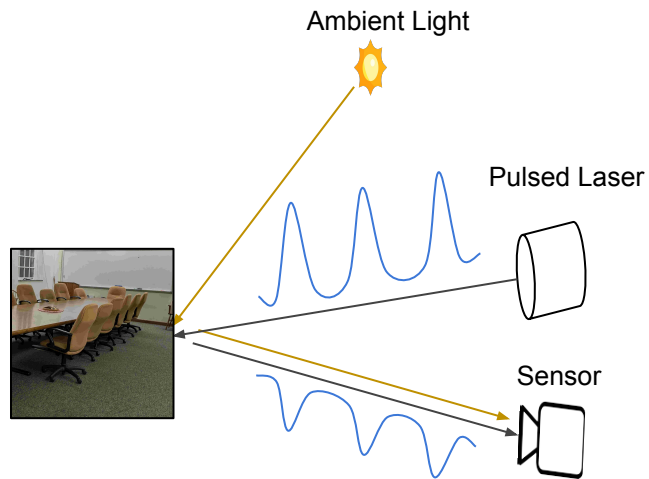
Timing Histogram



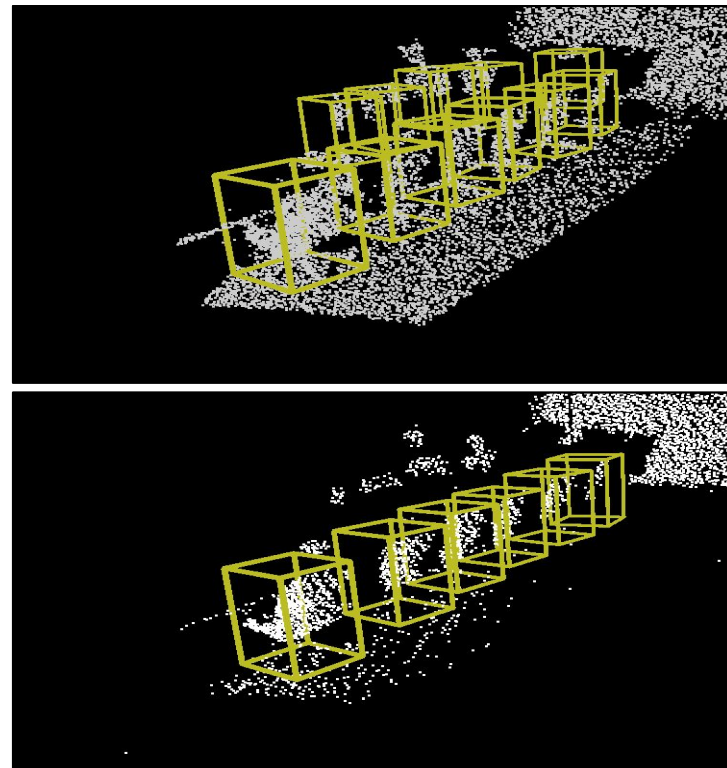
Point Cloud



# 3D Inference Pipeline with LiDAR



3D Point Cloud





# Challenging Low SBR in 3D Sensing

- Power constraints
- Lightweight sensors
- Low scene albedo
- Long distance
- Strong ambient light
- And more

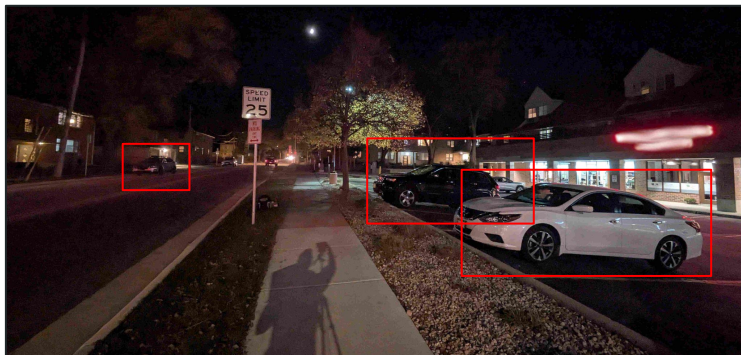


Resource Constrained Devices

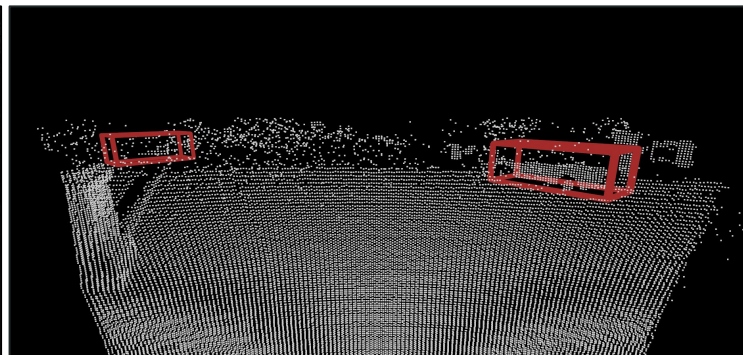
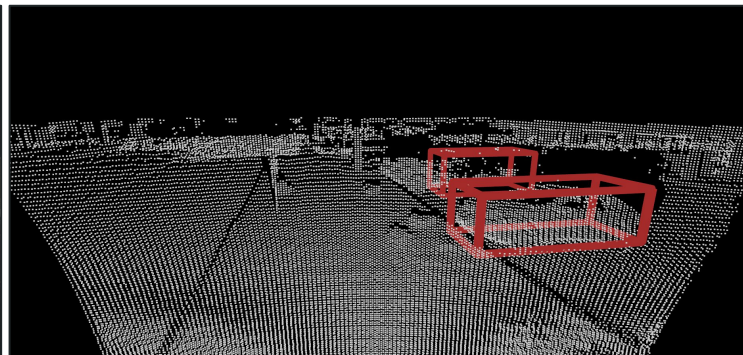


# 3D Inference in Low SBR

Scene (GT)

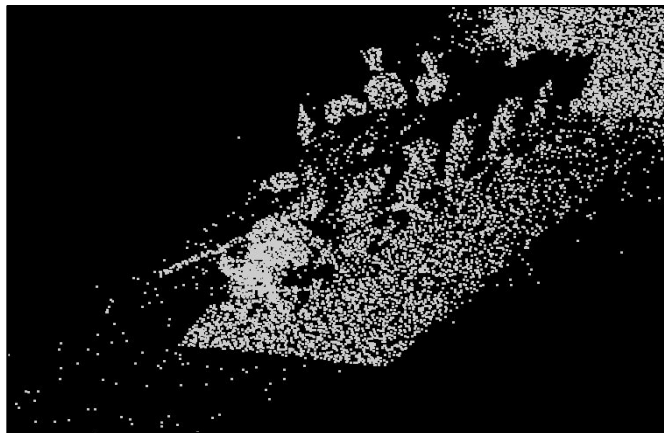


Point Cloud

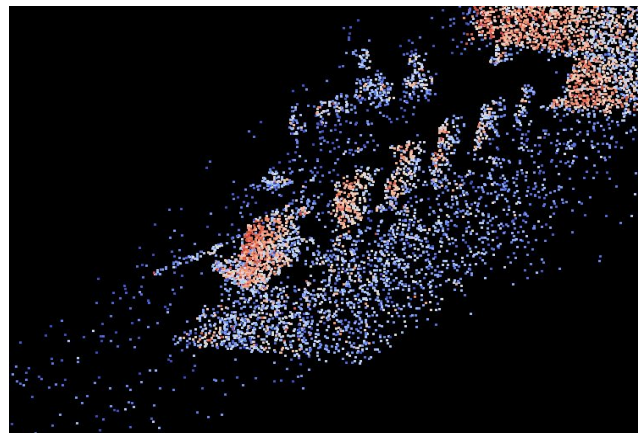
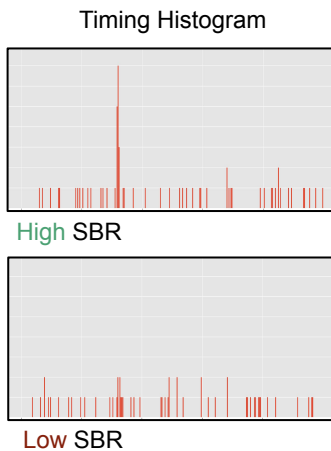


- Noisy and Sparse point clouds in Low SBR
  - Misses farther objects: cars

# Augmenting Point Cloud with Confidence



Point Cloud  
[x, y, z]

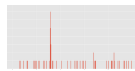
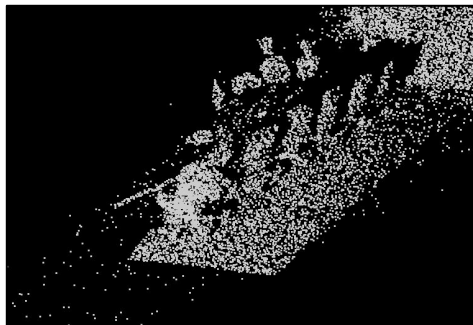


Probabilistic Point Cloud (PPC)  
[x, y, z, p]

$$Pr(p^{ij}) = \frac{h_{i,j}[m]}{\sum_{n=1}^N h_{i,j}[n]}, \quad \text{where } m = \arg \max_n h_{i,j}[n]$$

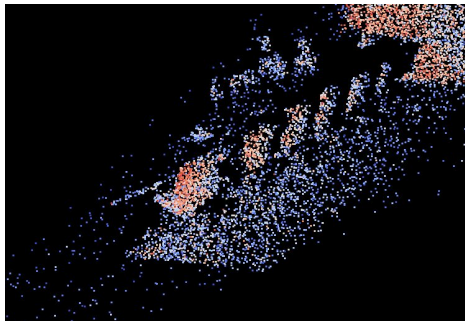
# Inference with Probabilistic Point Cloud (PPC)

Point Cloud

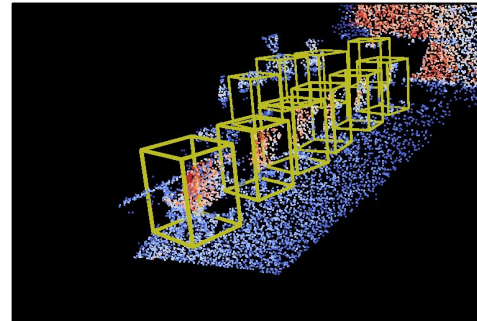


Probability  
Attribute

Probabilistic Point Cloud (PPC)



Robust  
Inference  
using PPC



- Probabilistic Point Cloud (PPC) from LiDARs
  - Robust Inference using PPC

# 3D Inference with PPC

- Neighbor Probability Density Filtering (NPD)
- Farthest Probable Point Sampling (FPPS)
- Probability Aware 3D inference networks

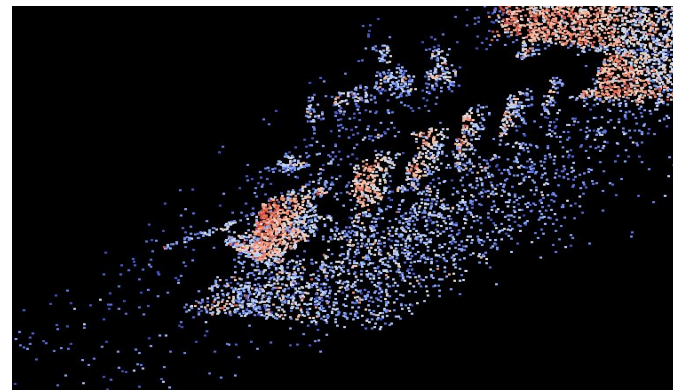
# 3D Inference with PPC

- Neighbor Probability Density Filtering (NPD)
- Farthest Probable Point Sampling (FPPS)
- Probability Aware 3D inference networks

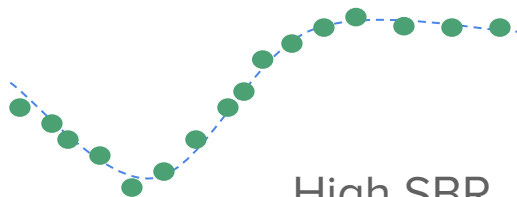


# Noise in Low SBR

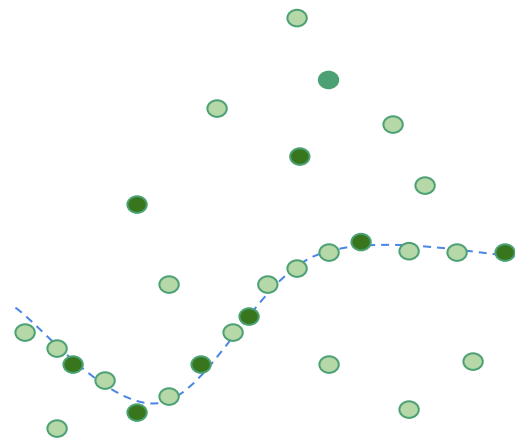
- Spurious points
  - Arbitrarily far from a surface/object
- High Spatial Density of points on surface



Probabilistic Point Cloud (PPC)



High SBR



Low SBR

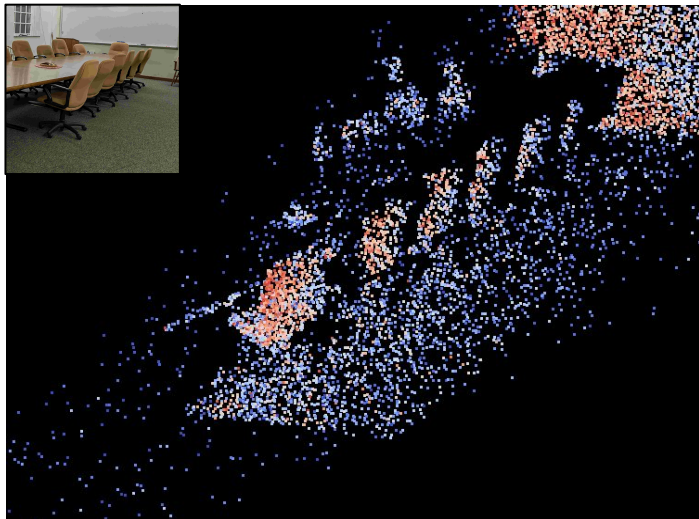
# Noise Removal

## Neighbor Probability Density (NPD) Filtering

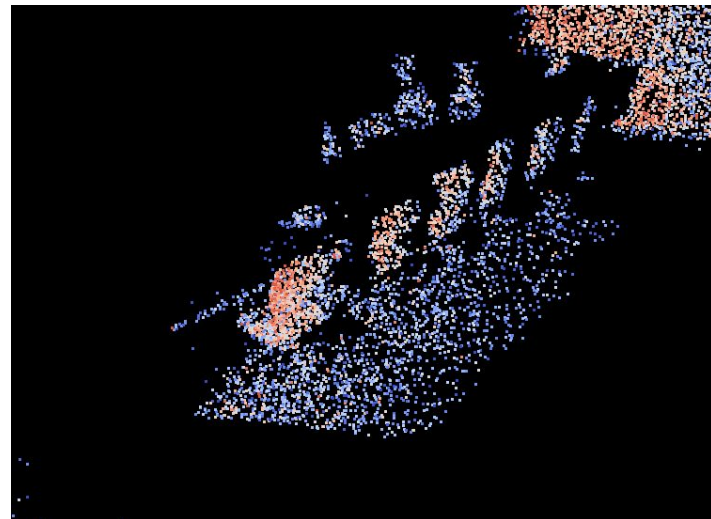
- High **Probability** and **Spatial Density**

$$NPD(p_i) = \frac{1}{L} * \sum_{p_j \in \mathcal{BQ}_{L,r}(p_i)} Pr(p_j)$$

NPD Score



NPD  
Filtering



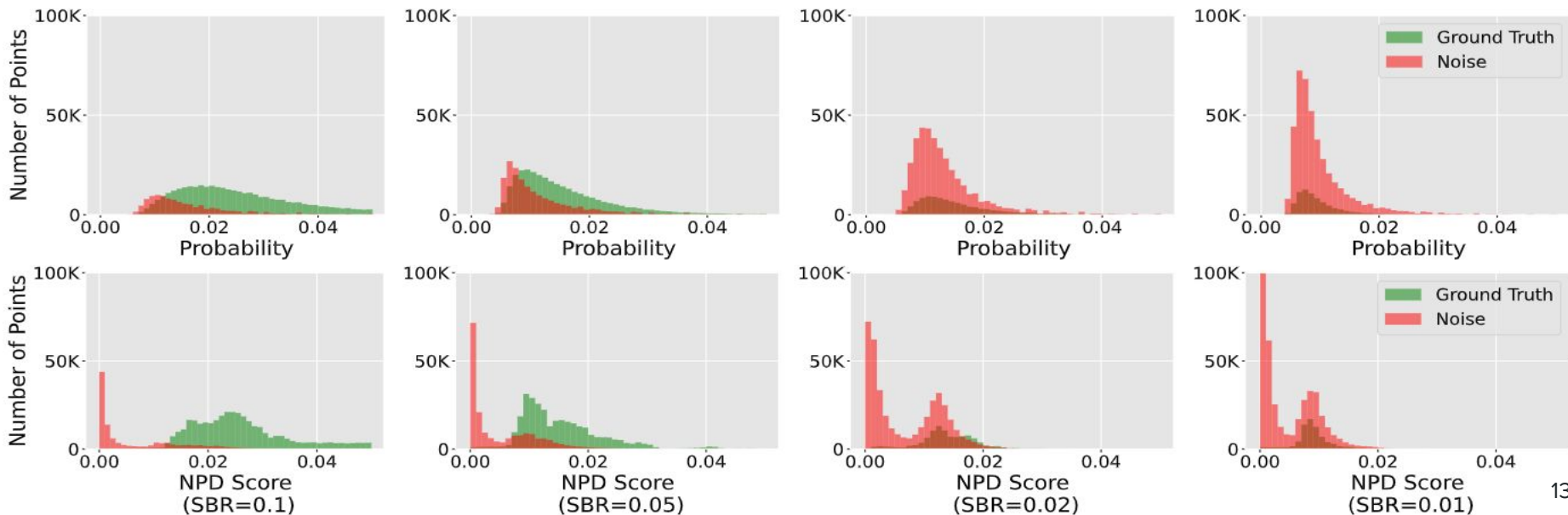
# Noise Removal

## Neighbor Probability Density (NPD) Filtering

- High **Probability** and **Spatial Density**

$$NPD(p_i) = \frac{1}{L} * \sum_{p_j \in \mathcal{BQ}_{L,r}(p_i)} Pr(p_j)$$

NPD Score

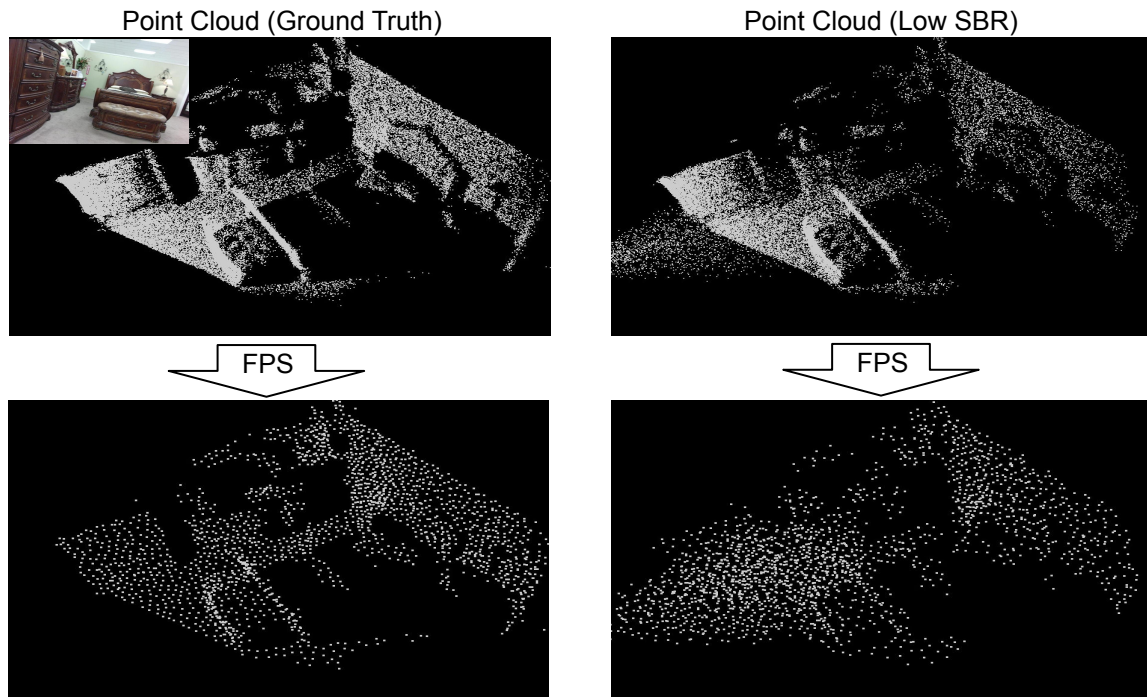


# 3D Inference with PPC

- Neighbor Probability Density Filtering (NPD)
- Farthest Probable Point Sampling (FPPS)
- Probability Aware 3D inference networks

# Point Sampling under Low SBR

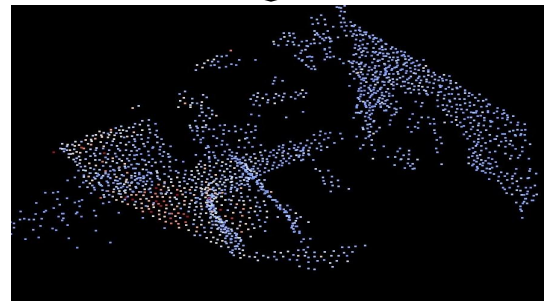
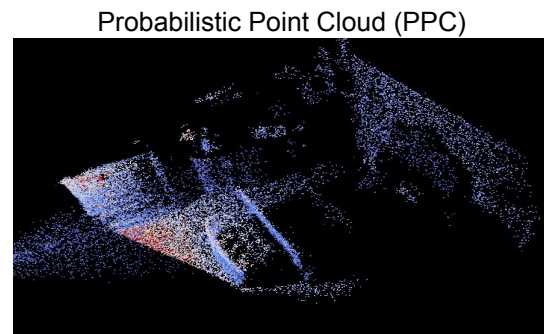
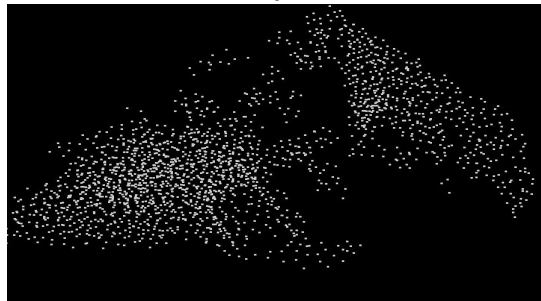
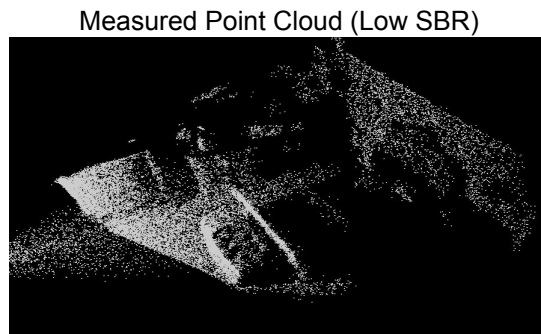
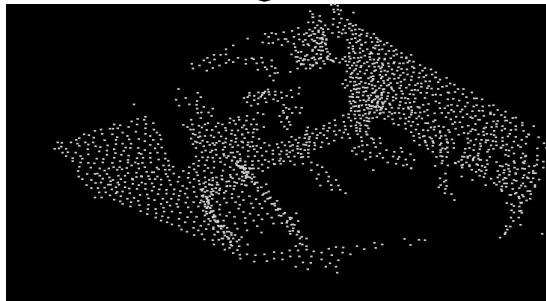
## Farthest Point Sampling (FPS)



FPS prioritizes farther points, samples **more noise**

# Farthest Probable Point Sampling (FPPS)

High probability candidate set for FPS



FPPS samples less noise

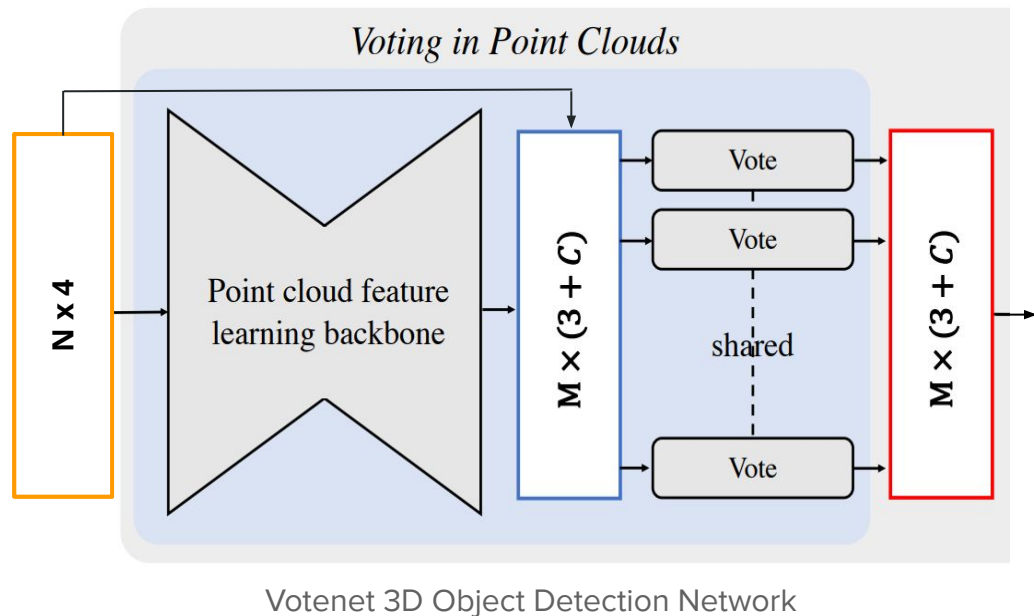


# 3D Inference with PPC

- Neighbor Probability Density Filtering (NPD)
- Farthest Probable Point Sampling (FPPS)
- Probability Aware 3D inference networks

# Probability Aware 3D Inference

- Probability as point attribute
- Probability weighted feature vectors



# 3D Object Detection Results

- LiDAR Setup and Real Captures
- Simulated Captures

# 3D Object Detection Results

- LiDAR Setup and Real Captures
- Simulated Captures

# LiDAR Setup



ADAPS LiDAR

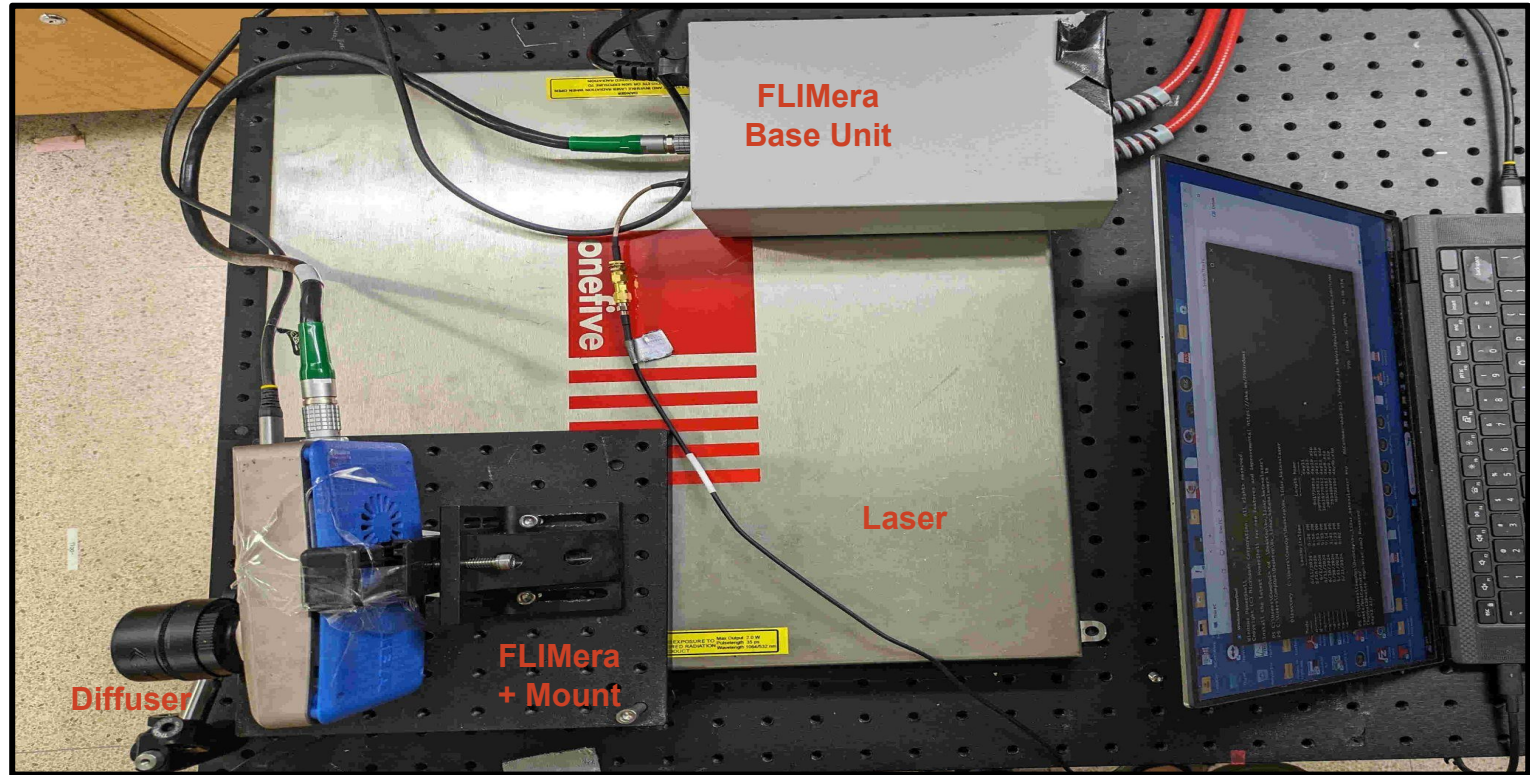
Chip Parameter	Spec
SPAD array	768 x 576
Macro pixel	3x3 SPAD binning
Macro pixel number	256 x 192
SPAD pitch	10um
SPAD Deadtime	~10ns
PDE (940nm)	>15%
Power Supply	VOP: -20V~30V
	AVCC: 3.3V
	DVCC: 1.1V
	IOVCC: 1.8V

Chip Parameter	Spec
Output mode	1) Grayscale mode
	2) Histogram mode
	3) Echo mode
	4) Ranging mode
Interface	MIPI CSI-2 DPHY(4 Lane) I2C/SPI
Max detection range	200 m
Chip size	< 100 mm <sup>2</sup>
Power consumption	~ 3W
TX channel	4
Package	iBGA/COB

System Parameter	Spec
Detection Range	30m
FOV (H)	120°
FOV (V)	90°
Angular resolution	0.5° x 0.5°

System Parameter	Spec
Frame rate	20 Hz
Accuracy	3cm
Laser wavelength	940nm
Eye safety	Class 1

# LiDAR Setup



FLIMera LiDAR



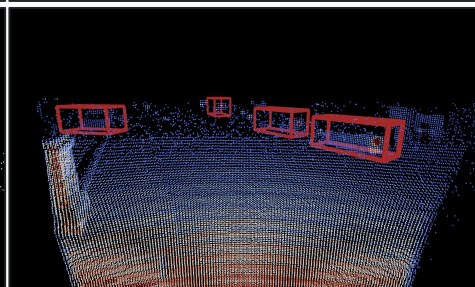
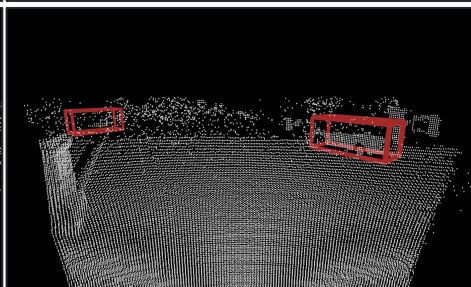
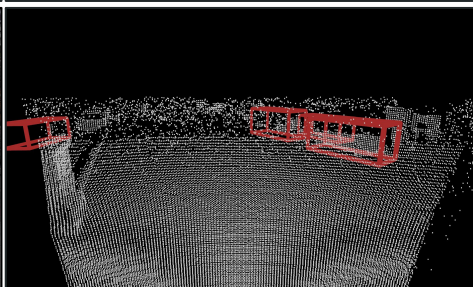
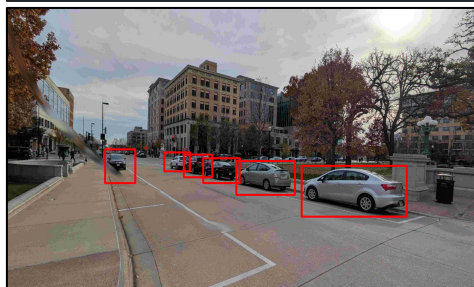
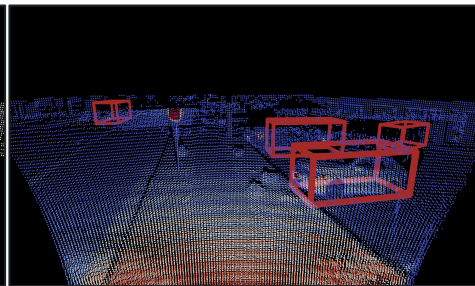
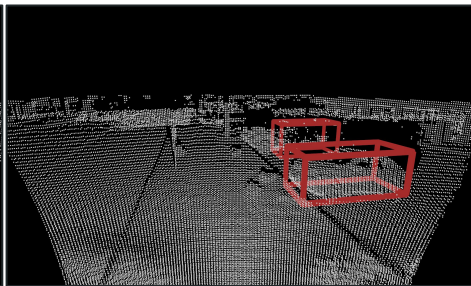
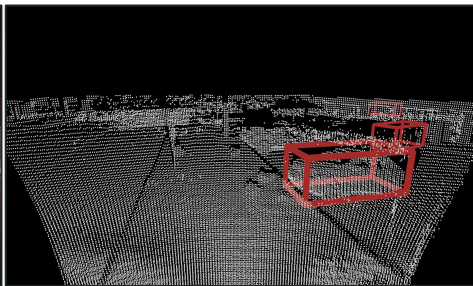
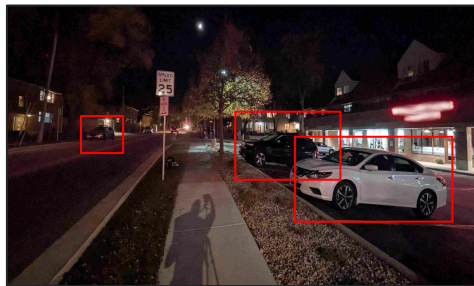
# 3D Object Detection on Real Outdoor Captures

Scene (GT)

Matched Filtering

Thresholding

PPC (Ours)



PPC detects all objects: farther **cars**

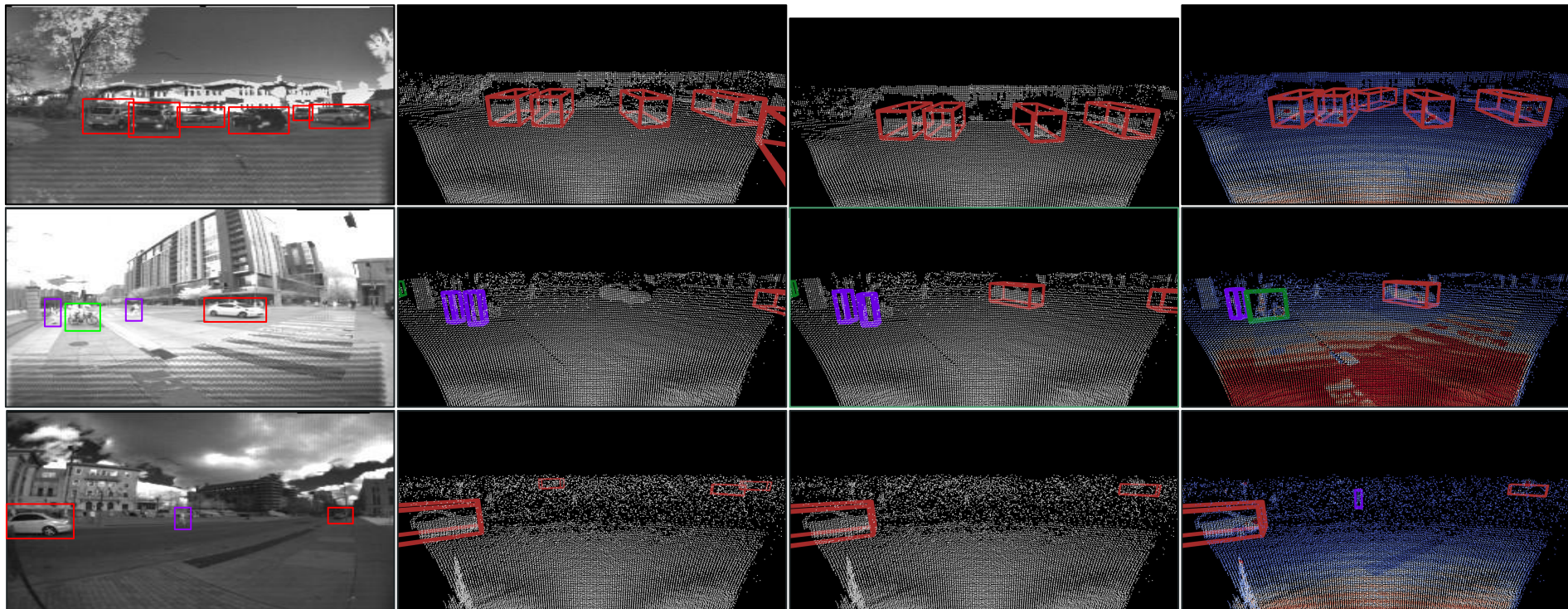
# 3D Object Detection on Real Outdoor Captures

Scene (GT)

Matched Filtering

Thresholding

PPC (Ours)



PPC detects all objects: far **car**, **pedestrian** and **bike**



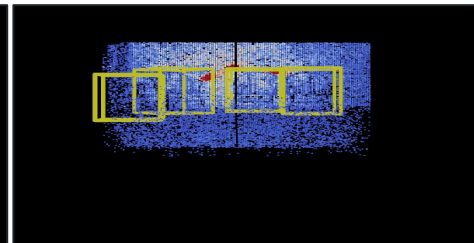
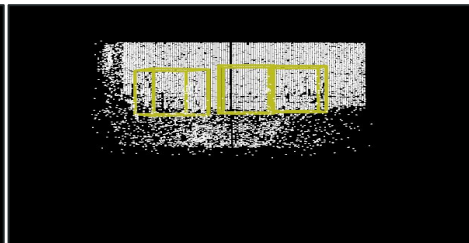
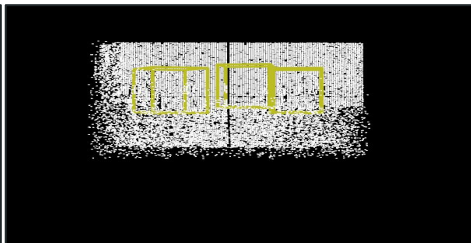
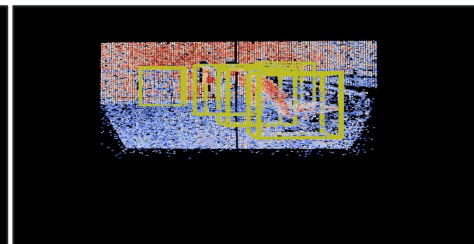
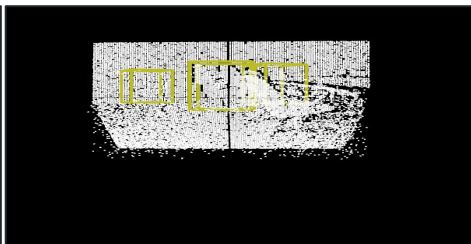
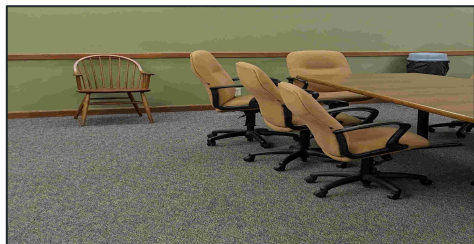
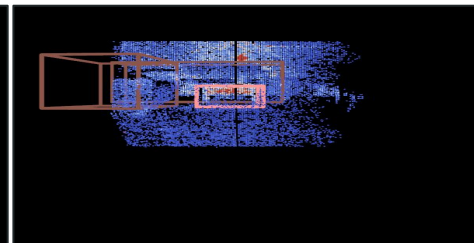
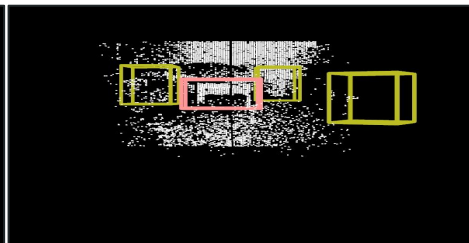
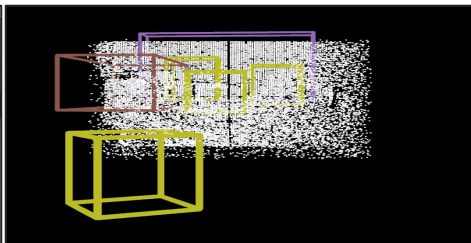
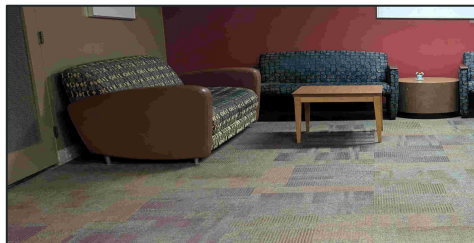
# 3D Object Detection on Real Indoor Captures

Scene (GT)

Matched Filtering

Thresholding

PPC (Ours)



PPC detects all objects: couch, far chair and table

# 3D Object Detection Results

- LiDAR Setup and Real Captures
- Simulated Captures

# 3D Detection on SUN RGB-D

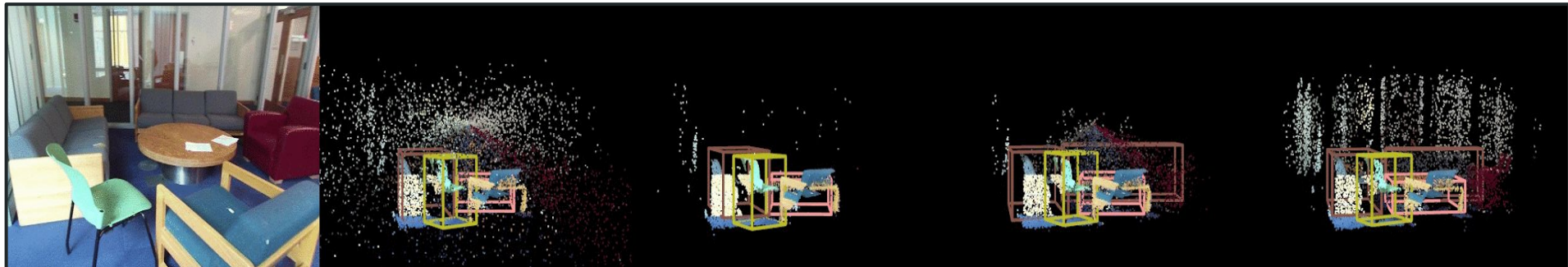
Scene

Matched Filtering

Thresholding

PPC (Ours)

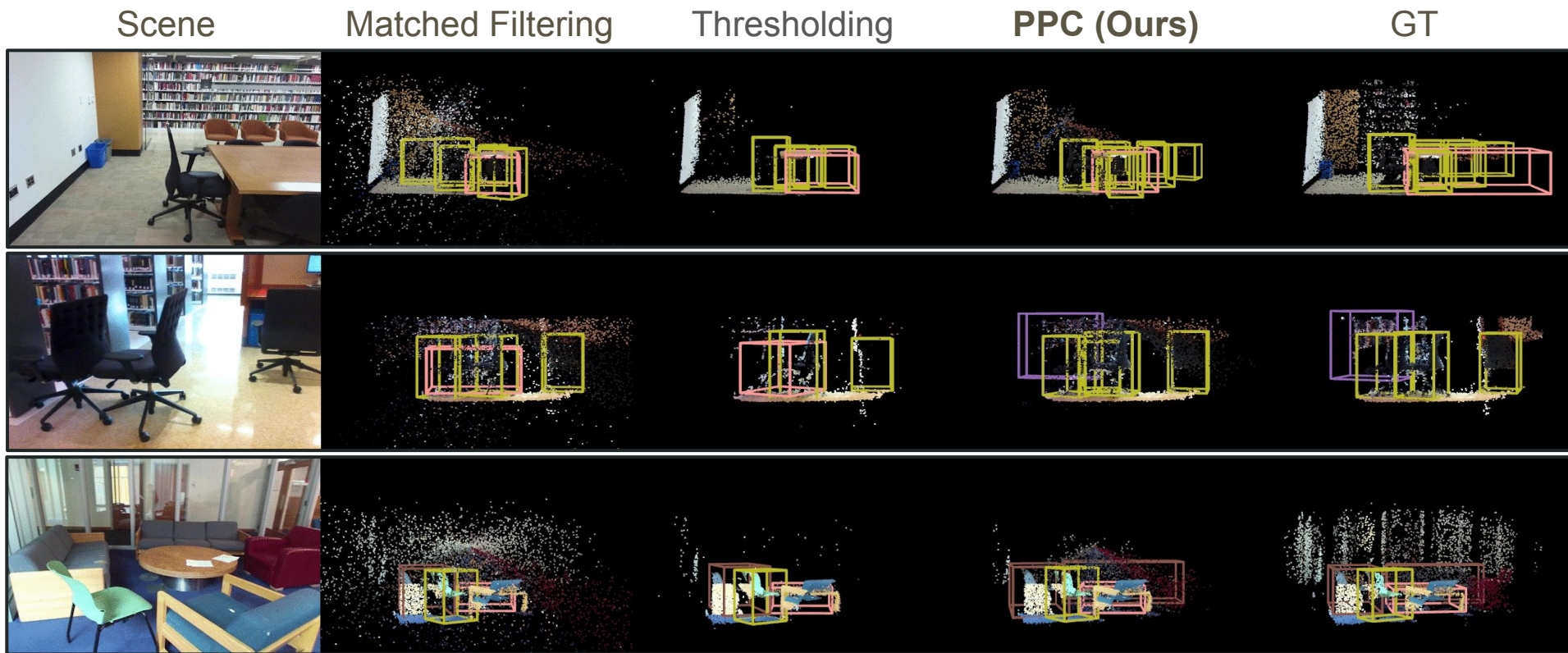
GT



Avg. SBR	Clean		0.1		0.05		0.02		0.01	
	AP@25	AP@50	AP@25	AP@50	AP@25	AP@50	AP@25	AP@50	AP@25	AP@50
Matched Filtering	51.34	27.45	42.43	20.49	38.77	17.57	16.95	5.05	11.34	2.73
Thresholding	57.11	33.21	51.27	28.62	46.44	24.86	<u>29.58</u>	<u>14.81</u>	<u>16.47</u>	<u>6.45</u>
PointClean Net [47]	54.58	31.89	45.65	26.44	40.19	19.15	18.24	8.05	12.78	3.01
Score Denoising [33]	<u>57.38</u>	<u>34.02</u>	<u>53.19</u>	<u>29.45</u>	<u>48.61</u>	<u>25.78</u>	26.35	13.73	14.55	4.73
PathNet [57]	57.16	33.87	52.16	28.79	47.11	24.89	25.45	12.96	13.87	4.56
<b>PPC (Ours)</b>	<b>58.61</b>	<b>34.99</b>	<b>54.29</b>	<b>31.15</b>	<b>52.46</b>	<b>30.20</b>	<b>38.49</b>	<b>16.47</b>	<b>29.42</b>	<b>13.16</b>

PPC outperforms all baselines significantly under very low SBR

# 3D Object Detection on SUN RGB-D



PPC can detect farther objects: **bookshelf**, **couch** and **chair**



# 3D Detection on KITTI

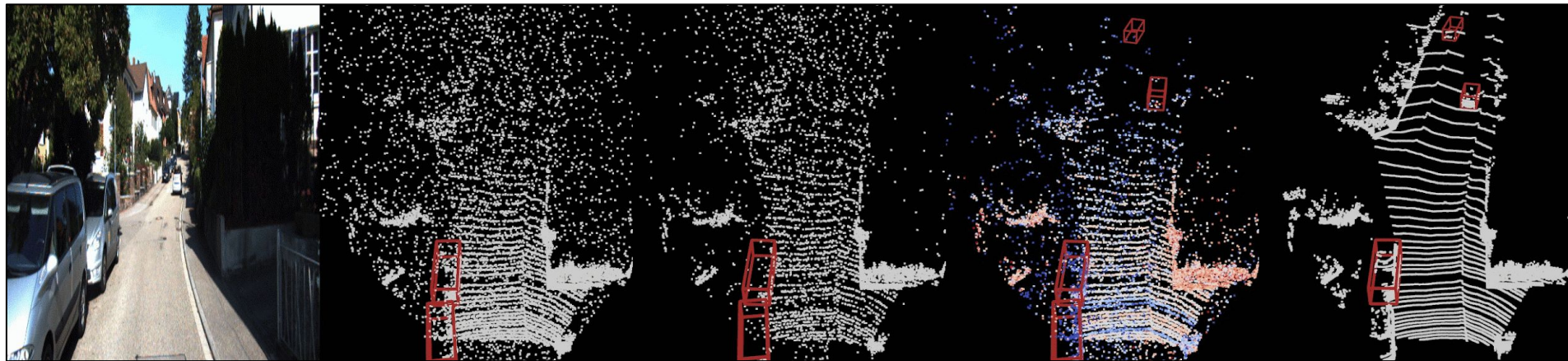
Scene

Matched Filtering

Thresholding

PPC (Ours)

GT



Avg. SBR	Clean			0.05			0.02			0.01			0.005		
	Car	Ped	Cyc	Car	Ped	Cyc	Car	Ped	Cyc	Car	Ped	Cyc	Car	Ped	Cyc
Matched Filtering	82.48	60.11	71.36	<b>73.14</b>	55.76	61.84	68.17	50.03	52.85	59.95	47.06	43.74	50.68	37.01	35.01
Thresholding	82.81	58.63	71.55	72.80	57.72	60.44	68.05	54.80	52.71	59.40	49.23	44.96	50.35	38.62	35.74
<b>PPC (Ours)</b>	<b>83.56</b>	<b>60.62</b>	<b>73.35</b>	73.03	<b>59.12</b>	<b>64.14</b>	<b>68.42</b>	<b>59.04</b>	<b>53.18</b>	<b>60.29</b>	<b>55.39</b>	<b>47.76</b>	<b>51.30</b>	<b>49.51</b>	<b>36.44</b>

PPC outperforms all baselines significantly under very low SBR

# 3D Object Detection on KITTI

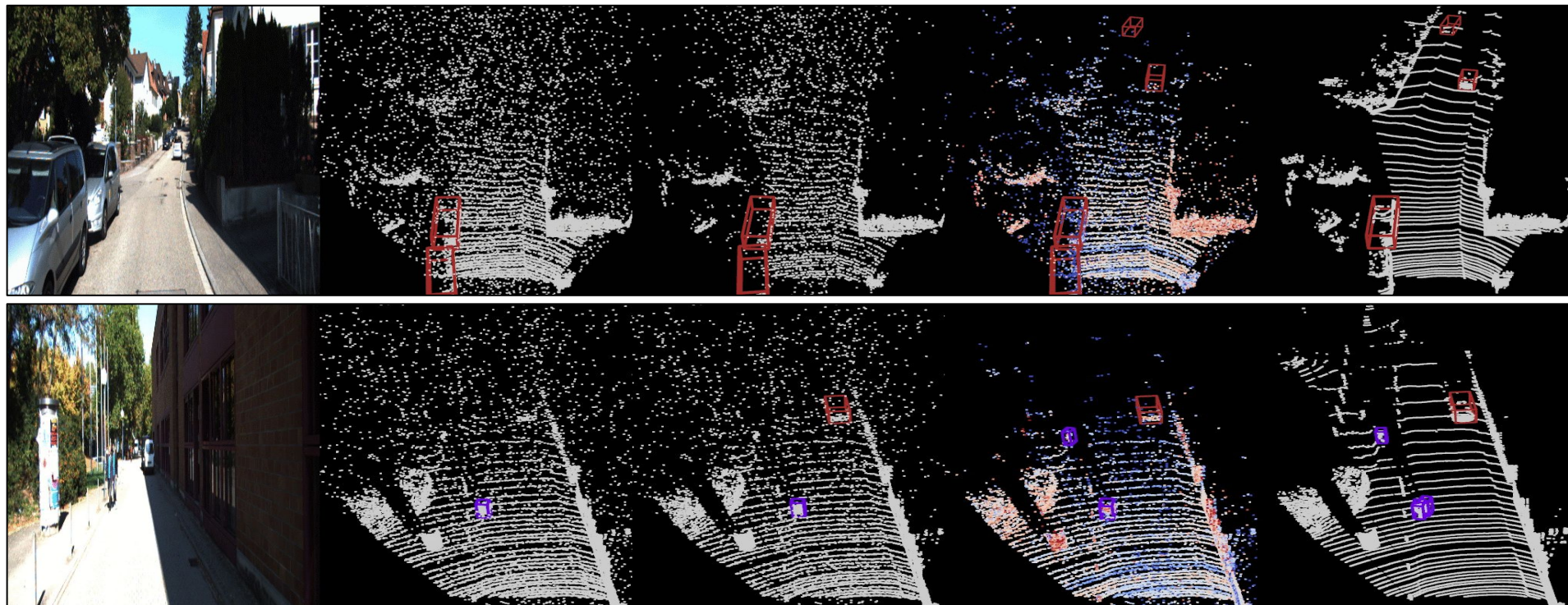
Scene

Matched Filtering

Thresholding

PPC (Ours)

GT



PPC can detect farther objects: **car** and far **pedestrian**



# 3D Object Detection on KITTI

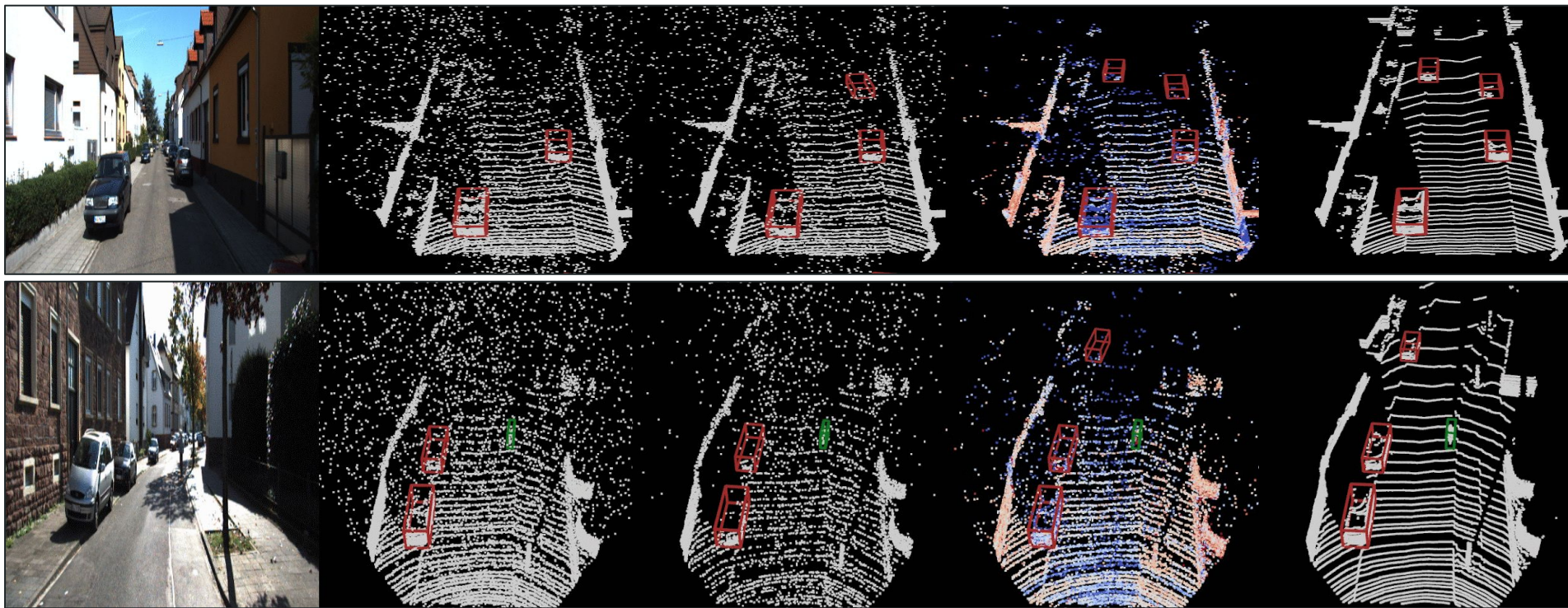
Scene

Matched Filtering

Thresholding

PPC (Ours)

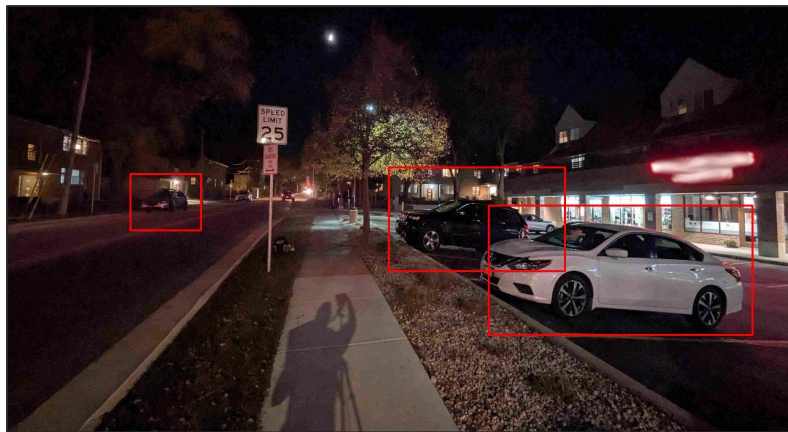
GT



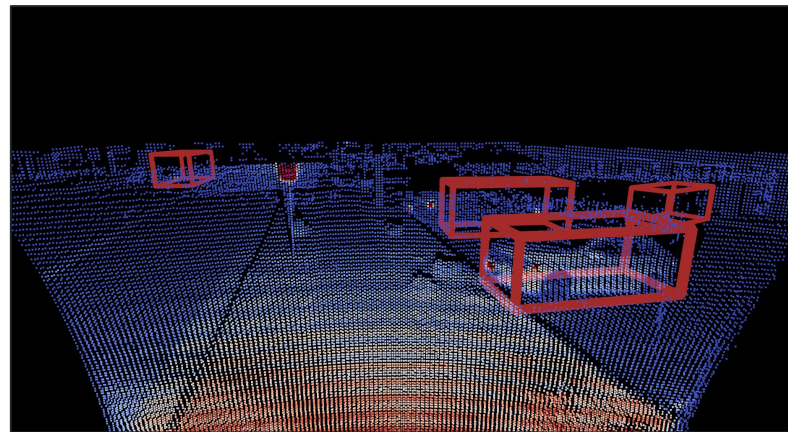
PPC can detect distant objects: cars

# Summary

Scene



Probabilistic Point Cloud



**Sensing:** Confidence information from 3D sensors

**Inference:** Robust 3D perception using probabilistic point clouds

# Thank You

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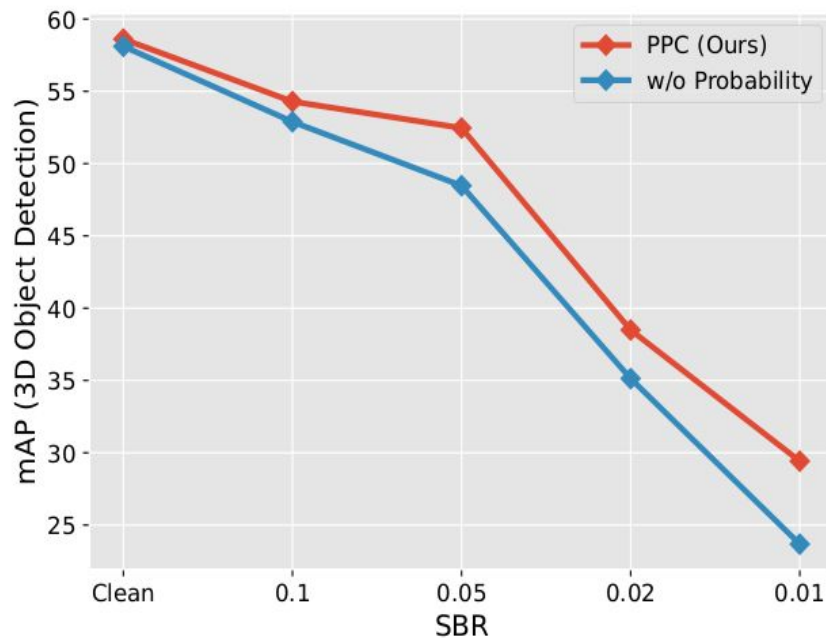
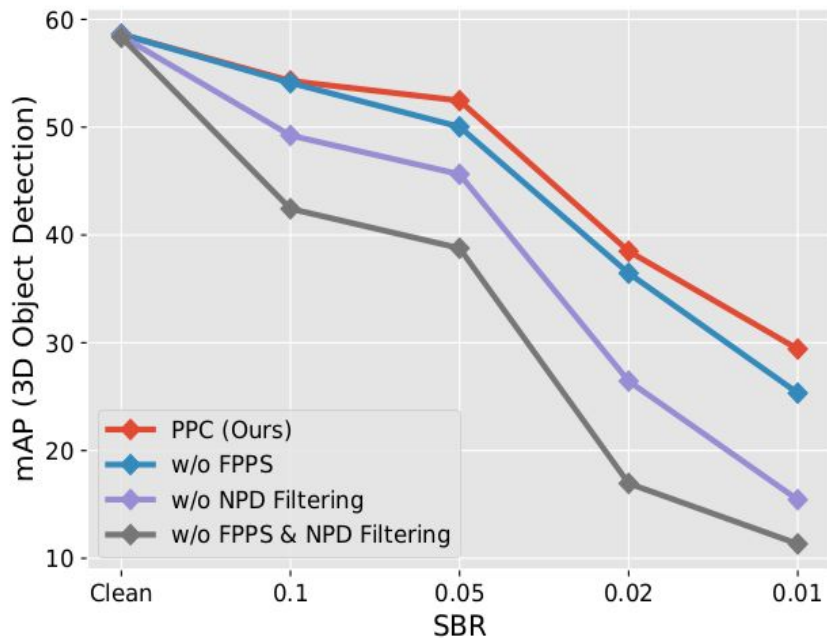
Questions?



# Additional Results

# Ablation Study

- Probability Attribute
- NPD Filtering and FPPS

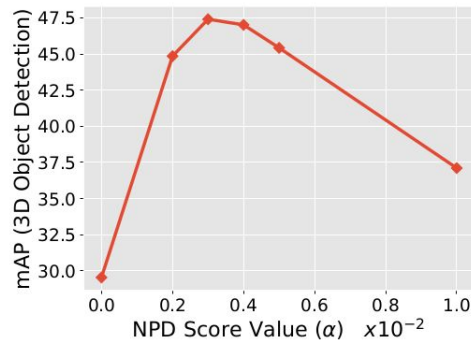




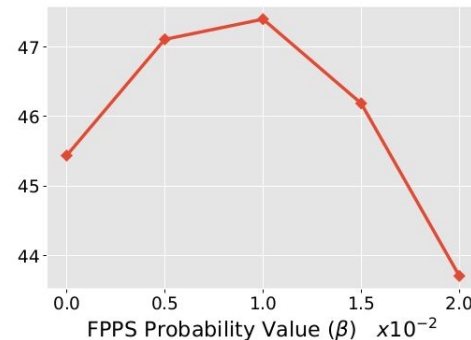
# Ablation Study

- NPD Score Hyperparameters

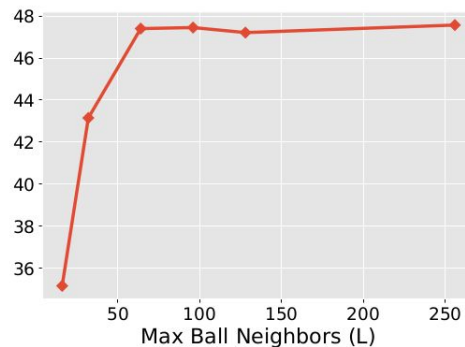
- NPD Score ( $\alpha$ )
- FPPS Probability ( $\beta$ )
- Max Ball Neighbors (L)
- Ball Radius (r)



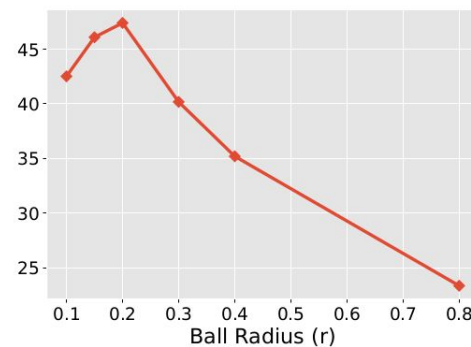
(a)



(b)



(c)



(d)

# Camera-LiDAR fusion with ImVoteNet

Avg. SBR	Clean		0.1		0.05		0.02		0.01	
	AP@25	AP@50	AP@25	AP@50	AP@25	AP@50	AP@25	AP@50	AP@25	AP@50
Matched Filtering	63.37	35.51	53.89	27.64	53.23	24.67	37.54	10.99	33.17	7.98
Thresholding	64.25	36.17	59.57	33.44	58.82	32.45	42.43	18.17	39.51	12.61
<b>PPC (Ours)</b>	<b>64.36</b>	<b>36.94</b>	<b>61.51</b>	<b>35.69</b>	<b>60.19</b>	<b>31.38</b>	<b>53.21</b>	<b>25.37</b>	<b>46.84</b>	<b>20.14</b>

PPC outperforms all baselines significantly on SUN RGB-D under very low SBR

# PointPillars Detector

Avg. SBR	Clean			0.05			0.02			0.01			0.005		
	Car	Ped	Cyc	Car	Ped	Cyc	Car	Ped	Cyc	Car	Ped	Cyc	Car	Ped	Cyc
Matched Filtering	77.08	<b>52.78</b>	64.49	68.25	49.52	58.96	64.13	47.67	46.45	54.43	41.61	41.76	45.03	32.46	31.06
Thresholding	<b>77.34</b>	52.09	64.81	68.06	49.63	59.09	63.87	47.92	46.96	54.18	40.88	42.18	45.11	32.79	31.89
<b>PPC (Ours)</b>	77.19	52.12	<b>65.21</b>	<b>69.12</b>	<b>50.23</b>	<b>62.44</b>	<b>65.63</b>	<b>49.27</b>	<b>48.09</b>	<b>56.39</b>	<b>45.77</b>	<b>44.46</b>	<b>47.24</b>	<b>38.74</b>	<b>34.89</b>

PPC outperforms all baselines significantly on KITTI under very low SBR

# Uni3DETR (Transformer-based) Detector

Avg. SBR	Clean		0.1		0.05		0.02		0.01	
	AP@25	AP@50	AP@25	AP@50	AP@25	AP@50	AP@25	AP@50	AP@25	AP@50
Matched Filter	64.98	48.28	61.52	45.09	60.82	43.92	51.12	34.09	45.29	27.97
Thresholding	64.50	47.94	61.08	44.71	61.19	44.29	51.97	34.87	48.13	28.64
<b>PPC (Ours)</b>	<b>65.53</b>	<b>49.35</b>	<b>62.58</b>	<b>46.71</b>	<b>61.98</b>	<b>48.28</b>	<b>56.46</b>	<b>38.03</b>	<b>51.21</b>	<b>31.16</b>

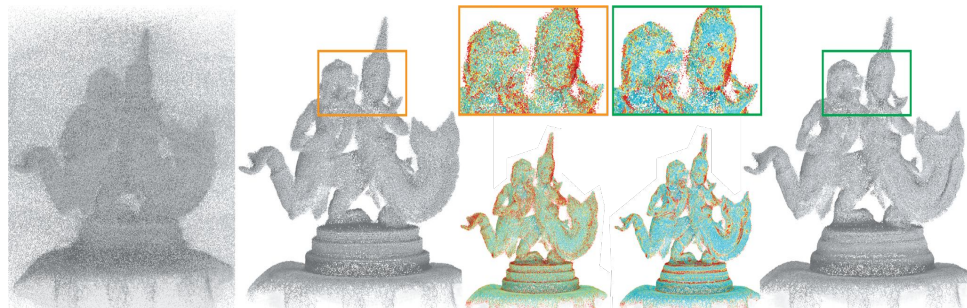
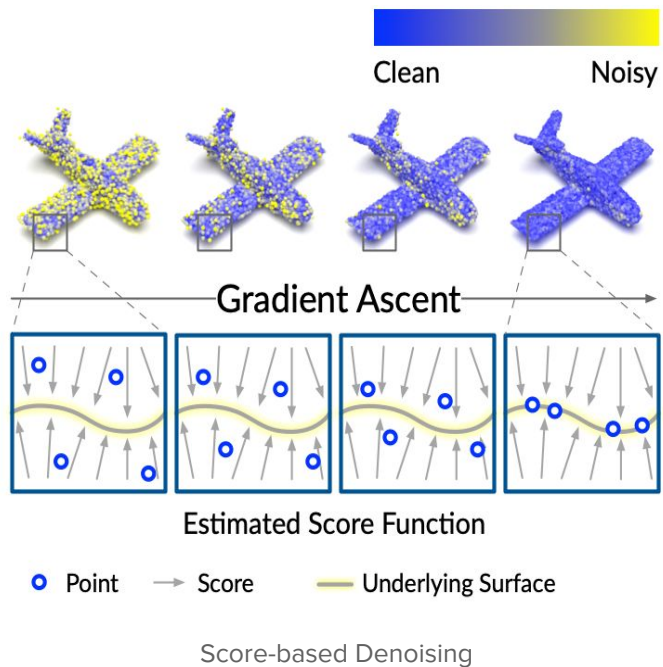
PPC outperforms all baselines significantly on SUN RGB-D under very **low SBR**

# Category-wise Results on SUN RGB-D

Category	Matched Filtering	Thresholding	PointClean Net [41]	Score Denoise [28]	PPC (Ours)
Bed	54.37	67.33	53.59	68.57	<b>72.97</b>
Sofa	16.20	28.15	22.00	38.25	<b>45.19</b>
Table	29.61	37.99	30.20	33.93	<b>40.47</b>
Bathtub	6.04	25.14	2.71	14.04	<b>54.32</b>
Desk	7.97	14.16	7.78	8.33	<b>17.37</b>
Bookshelf	2.12	4.85	0.67	1.01	<b>9.80</b>
Chair	22.05	34.27	22.92	27.45	<b>47.47</b>
Night Stand	3.10	14.45	7.41	10.23	<b>30.49</b>
Dresser	2.58	4.64	2.80	2.44	<b>13.73</b>

PPC is very effective on small categories under very low SBR

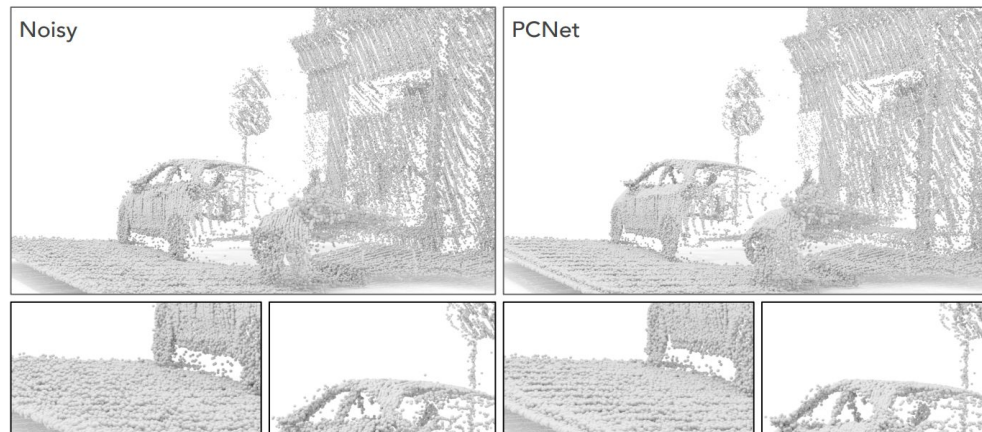
# Denoising Networks



PointClean Net

# Should we use denoising networks for inference?

- Improves Surface Reconstruction
  - Not 3D Object Detection
- Local Isotropic Gaussian Noise
- Spurious point have no depth information



- Computational Overhead

	Matched Filtering	Thresh- olding	PointClean Net	Score- Denoising	PathNet	PPC (Ours)
Runtime (ms)	87	89	755	1345	867	95



# Runtime

	Matched Filtering	Thresh- olding	PointClean Net	Score- Denoising	PathNet	PPC (Ours)
Runtime (ms)	87	89	755	1345	867	95

Comparison of inference time

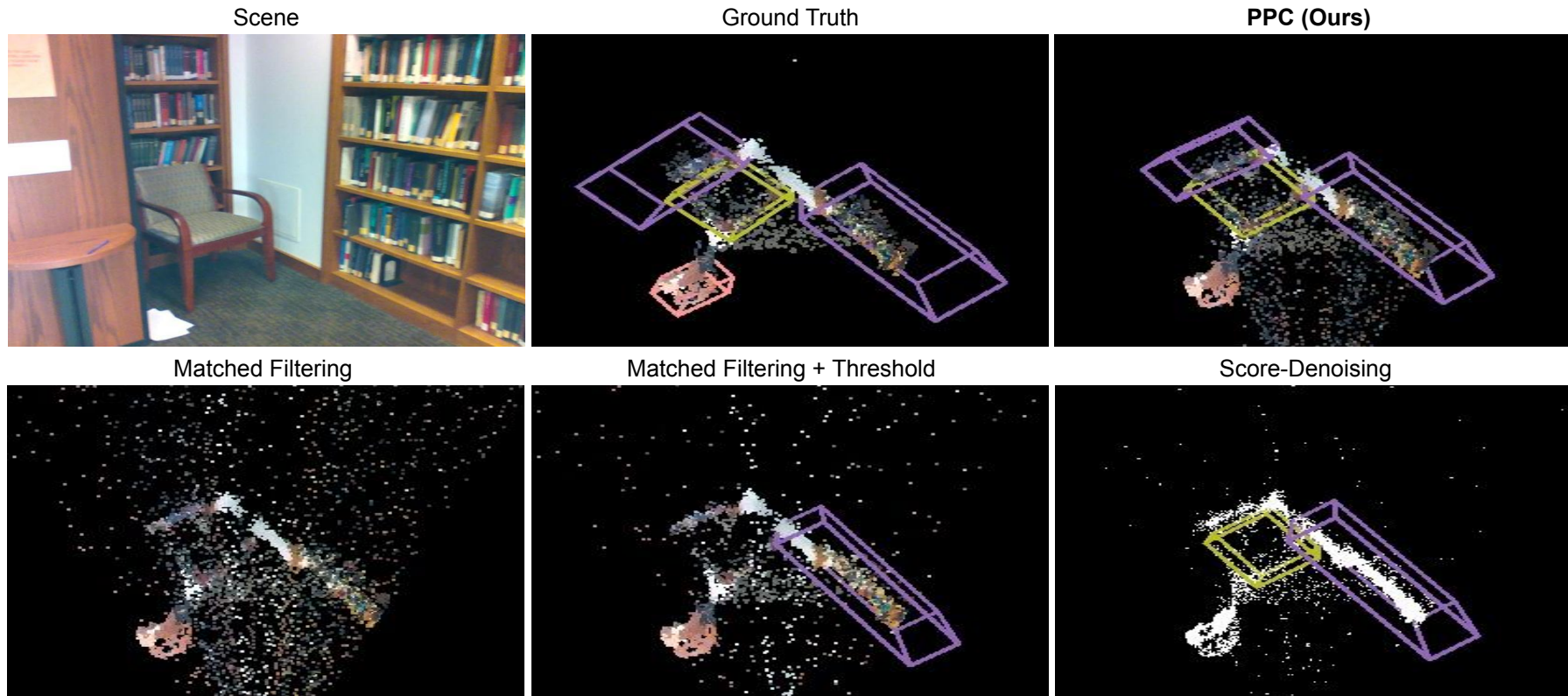
## 3D Detection Results with PPC Variants

PPC-aware detection models show better performance

Avg. SBR	Clean	0.1	0.05	0.02	0.01
Thresholding	57.11	51.27	46.44	29.58	16.47
PPC	58.61	54.29	52.46	38.49	29.42
PPC w/ prob point att	58.35	55.73	<u>53.67</u>	<u>39.10</u>	<u>30.58</u>
PPC w/ wt. point feat vec	<u>59.29</u>	<u>55.86</u>	<u>52.95</u>	37.68	30.42
PPC w/ wt. obj score	58.53	55.18	53.15	38.76	30.17
PPC w/ all	<b>59.35</b>	<b>56.11</b>	<b>53.45</b>	<b>39.87</b>	<b>30.81</b>

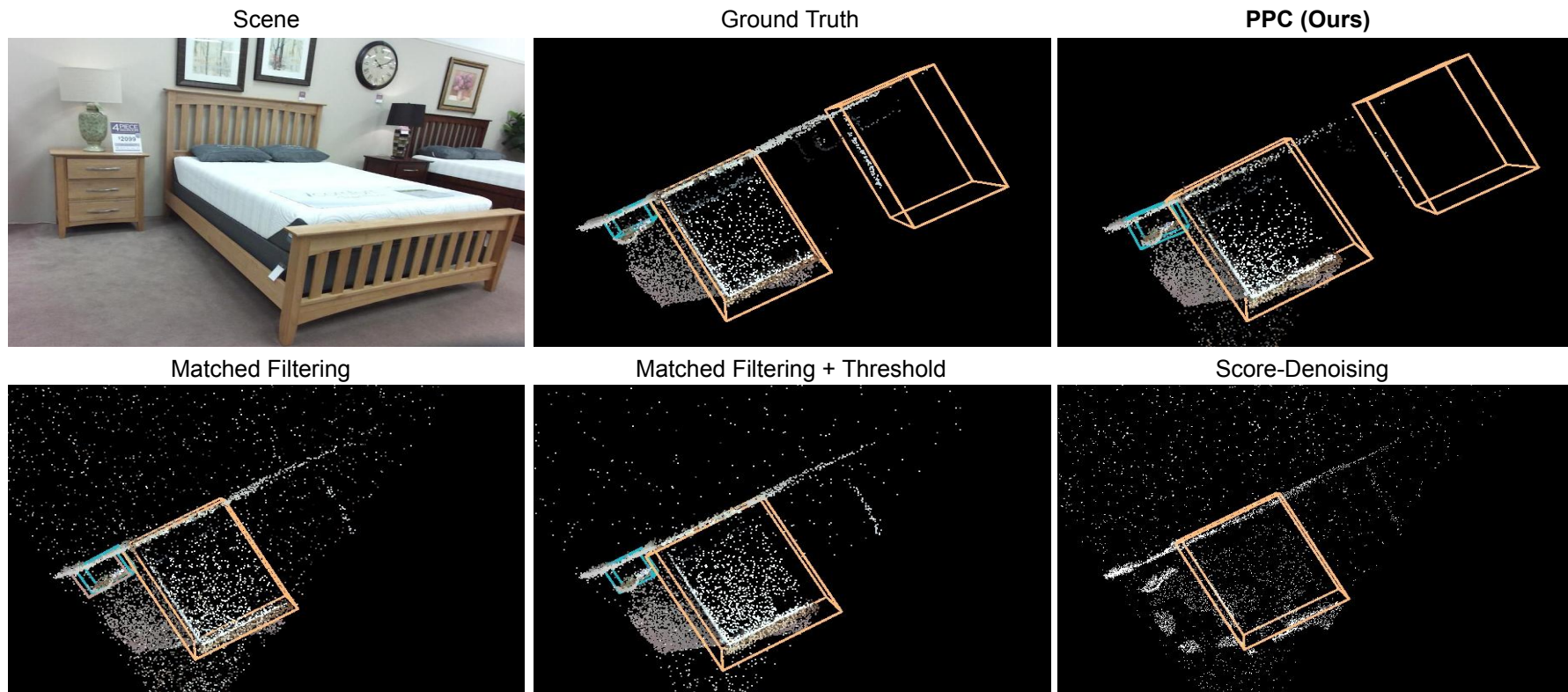
Table 5. AP@25 results on SUN RGB-D with PPC variants.

# 3D Detection Results



PPC can detect **smaller** and **farther** objects

# 3D Detection Results



PPC detects **smaller** and **farther** objects